

ENHANCEMENT OF SEMANTIC MATCHMAKING IN WEB SERVICES

SAVAN M. PATEL¹ & GAURAV K. PATEL²

¹Department of CE, U. V. Patel Engineering College, Kherva, Gujarat, India

²Department of CSE, LCIT College, Bhandu, Gujarat, India

ABSTRACT

The ability to dynamically discover and invoke a Web Service is a critical aspect of Service Oriented Architectures. An important component of the discovery process is the matchmaking algorithm itself. In order to overcome the limitations of a syntax-based search, matchmaking algorithms based on semantic techniques have been proposed. As a greater number of Web Services are made available, support for service discovery mechanisms become essential. So we have use the semantic matchmaking using the greedy approach, that work on the greedy concept. The reault shows that the matchmaking with greedy concept gets the fast requested service as compare to the other algorithms. Matchmaking plays a vital role in this discovery process.

KEYWORDS: Web Service Composition, Ontology, Web Services, Semantic Web Services. Semantic Similarity

INTRODUCTION

With contributions from Semantic Web technology, Web services are enhanced using rich description languages such as Web Ontology Language [1]. Semantic Web Services [2] are Web services that have been enhanced with formal semantic descriptions in Web Ontology Language of Web services [3]. OWL-S is an OWL-based Web services ontology that provides Web service suppliers with a core set of markup languages that is constructed for describing properties and functions, and the grounding information for Web services in a computer-interpretable form.

As each service is added to constructed composition workflow, a parameter matching mechanism is necessary to make sure that the service supports the **Inputs, Outputs, Preconditions, and Effects (I/O/P/E)** constraints of the workflow. In addition, I/O/P/E parameters of the selected services affect the matching step for the next node of the constituted workflow. Therefore, in order to obtain proper compositions of Web services, a powerful matching process is required. Including potential services in a composition task is critical for the assessment of the composition.

The Semantic Web approaches to web services give us the ability to describe the semantics of web services and their capabilities in a formal and machine-process manner. The Semantic Web should enable greater access not only to content but also to services on the web. Users and software agents should be able to discover, invoke, compose, and monitor web resources offering particular services and having particular properties, and should be able to do so with a high degree of automation if desired. The focus of this paper is on using the semantic meta data which is published for web services for measuring the similarity of web services. The ability of measuring the similarity of web services has several use cases. The following is a brief description of some of these use cases:

- **Service Discovery:** Semantic web service discovery or matchmaking is an emerging research area that exploits

the service semantic metadata to locate services that can perform tasks of given description with the best over all degree of satisfaction. Most of the proposed service discovery algorithms are based on measuring the similarity of the requested service with each of the advertised services. Therefore having a similarity measure for web services can be quite helpful in service discovery.

- **Service Recommendation:** It is often happens that a client spent some time for manually finding a service that can fulfil his requirements by browsing categories in a registry and then try to invoke the service but the service is unavailable. In this situation, similar services can be suggested to the client.
- **Service Composition:** Currently, many services are offered on the web and their number would increase exponentially in the future. But these atomic services may not be able to satisfy specific requests individually. So there must be brokers between service providers and service requesters which integrate the atomic services of service providers and create new value added services. The act of taking several services and bundling them together to meet the needs of a given customer is called service composition.

BACKGROUND AND RELATED WORK

Ontologies are used in order to incorporate semantics in web service descriptions. An Ontology models domain knowledge in terms of Concepts and Relationships between them. OWL^{[12][11]} has evolved as a standard for representation of ontologies on the Web. OWL-S^{[18][11]}, formerly called DAML-S^[10], defines an ontology for semantic web services. OWL-S describes a service in terms of its Service Profile, Process Model and Grounding. The Service Profile models the Inputs, Outputs, Pre-conditions and Effects (IOPE) of the service. The Inputs and Outputs in the Service Profile refer to concepts in ontologies published on the Web. Service Advertisements and search Queries are both expressed in terms of OWL-S descriptions.

An ontology reasoner is an important component in the process of semantic matchmaking. A reasoner can infer additional information which has not been explicitly stated in an ontology. Subsumption, concept satisfiability, equivalence and disjointness are some examples of reasoning operations. Many of these operations are used in the semantic matchmaking process. DAML-S is based on a logic formalism called Description Logics (DL). Description Logics and its reasoning are explored in detail by^[15] and^[19]. Racer^[7] and Pellet^{[5][23]} are some implementations of DL-Reasoners. The Service Profile contains enough information for a matchmaker to determine if a service satisfies the requirements of a client. In fact, several matchmaking algorithms rely only on the matching of Inputs and Outputs of the Service Profiles.

SEMANTIC MATCHMAKING ALGORITHM

The algorithm takes a OWLS *Query* from the client as input and iterates over every OWL-S *Advertisement* in its repository in order to determine a match. An *Advertisement* and a *Query* match if their Outputs and Inputs, both, match. The algorithm returns a set of matching advertisements sorted according to the degree of match.

Let Queryout and Advtout represent the *list of output concepts* of the *Query* and *Advertisement* respectively. Matching the outputs requires the matching between two *concept-lists*, Queryout and Advtout, as follows:

$$\forall c \in \text{Queryout}, \exists d \in \text{Advtout},$$

$$\text{s.t. match}(c, d) \neq \text{Fail}$$

Let $Query_{in}$ and $Advt_{in}$ represent the list of input Concepts of the Query and Advertisement respectively. Matching the inputs requires:

$$\forall c \in Advt_{in}, \exists d \in Query_{in},$$

s.t. $match(c, d) \neq Fail$.

Note that the order of *Query* and *Advt* has been transposed between the two expressions above. Suppose $outQ \in Query_{out}$ and $outA \in Advt_{out}$ are two concepts. In case of output matching, the $match(outQ, outA)$ function accepts $outQ$ and $outA$ as inputs and returns the degree of match between them. Four degrees of match are defined between a them:

- **Exact:** If $outA$ is an equivalent concept to $outQ$ or $outA$ is a superclass of $outQ$. In case of a superclass relationship, it is assumed that the service provider has agreed to support *every* possible subclass of $outA$.
- **Plugin:** If $outA$ Subsumes $outQ$. The relation between $outA$ and $outR$ is weaker as compared to the previous case since subsumption is indirectly inferred by the reasoner. It is assumed that the provider has agreed to support *some* sub-concepts of $outA$. We hence infer that $outA$ can be *plugged in place of* the required $outR$.
- **Subsume:** If $outQ$ Subsumes $outA$. The set of individuals defined by the concept, $outA$, is a subset of the set of individuals defined by the concept $outR$.
- **Fail:** If none of the above conditions are satisfied.

These four degrees are ranked as: $Exact > Plugin > Subsumes > Fail$. Here, $x > y$ indicates that x is ranked higher (is a more desirable match) than y .

This matching condition can be included in the some existing algorithms that gives the better results so we have decided to use the best predicting algorithm that is the greedy approach for enhancing the semantic web services. The greedy approach is detailed below.

Greedy Approach: The algorithm adopts a greedy approach towards matching the concept-lists. For example, in the case of output matching, for each concept in the $Query_{out}$, it determines a corresponding concept in $Advt_{out}$ to which it has a *maximum* degree of match. Once all such max-matchings are computed, the minimum match amongst them is the *overall degree of match* between the *Query* and the *Advertisement*.

PROBLEM FORMULATION AND ALGORITHM

Following are some definitions that are define in our approach for formulating the composition problem:

- **Domain Ontology (O):** A Domain Ontology O consists of a set of concepts used to describe the domain in which a group of services is inserted. These concepts are related to each other according to subsumption relation, i.e. a concept $c_1 \in O$ subsumes another concept $c_2 \in O$ if c_1 is a superclass of c_2 ; c_2 subsumes c_1 if c_1 is a subclass of c_2 ; c_1 and c_2 subsumes each other if they are the same.
- **Web Service (S):** A web service S is defined as $\{S_i, S_o\}$, where $S_i = \{S_1, S_2, \dots, S_i\}$ is a finite set of required input concept and $S_o = \{S_1, S_2, \dots, S_j\}$ is a finite set of provided output concept.
- **Context Parameter (C_p):** Context parameter C_p is variable that is attached with web service. We use user context to fulfil its requirement.

- **Request (R):** A composition request R consists of a set of provided inputs R_{in} , a set of required outputs R_{out} , and a context parameter C_p .
- **Service Repository (D):** A service repository D consists of a set of available services. Each service $S \in D$ is composed of a required inputs $S_i \subseteq O$, a set of provided outputs $S_o \subseteq O$, and a user context parameter C_p .

The Relationship between concepts of determined subsumption reasoning. So, given a required concept ($outR$) and an advertised concept ($outA$), the relationship between them can be labelled as follows

- **Exact:** If $outA$ and $outR$ are exactly the same concepts;
- **Plug-In:** If $outR$ is subsumed by $outA$ i.e. $outR$ is subclass of $outA$;
- **Subsumes:** If $outR$ subsumes $outA$ i.e. $outR$ is superclass of $outA$;
- **Fail:** Failure occurs when there is no subsumption relation between request and advertisement.

Subsumes: Two concepts c_1 & $c_2 \in O$ can be related to each other in four ways. We define predicate *subsumes*: $(O \times O) \rightarrow \{true, false\}$ to express this relation as follows:

- $subsumes(c_1, c_2)$ holds if only if c_1 is generalization of c_2 . (c_2 is then specialization of c_1)
- $subsumes(c_2, c_1)$ holds if only if c_2 is generalization of c_1 . (c_1 is then specialization of c_2)
- if neither $subsumes(c_1, c_2)$ nor $subsumes(c_2, c_1)$ holds, c_1 and c_2 are not related to each other.
- $subsumes(c_1, c_2)$ and $subsumes(c_2, c_1)$ is true if and only if $c_1 = c_2$.

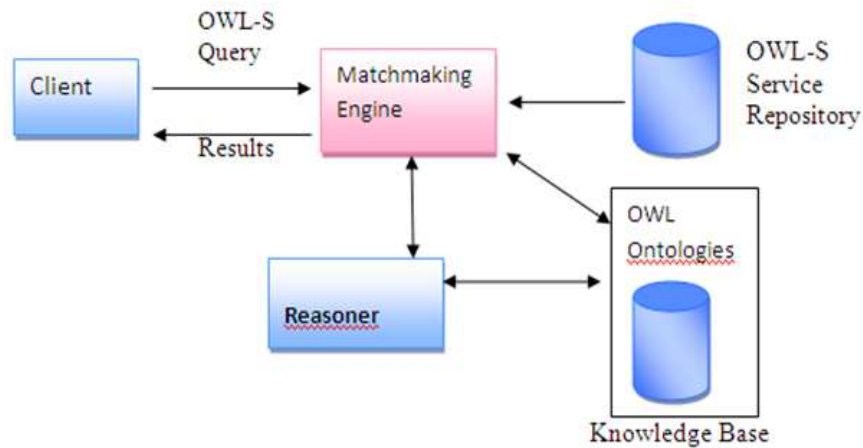


Figure 1: Proposed Framework

The operation of the framework can be described as follows:

A composition process begins when a Service Requester requests some functionality from service. Service Requester provides input parameters of service, required output and Quality of Service constraints as part of composition request.

Composition Engine starts to discover a set of services those provides required output and fulfils QoS constraints imposed by Requester from Service Repository with the help of Matchmaker.

Matchmaker module performs matchmaking operation between User's Query and Services provided by Repository. Reasoner is used with Matchmaker for computing similarity measure of the concepts. Matchmaker provides matched services to the Service Selection & Filtering module for selection, sorting and filtering of candidate services for composition. QoS parameters i.e. Throughput and Response Time are utilized for filtering purpose. The matchmaking process is performed with the help of domain ontology.

COMPOSITION ALGORITHM

In this section we propose algorithms for Semantic Matchmaking of Web services.

Algorithm 1

Following algorithm defines the matchmaking process by considering the user request and available services from the repository.

Input: $R \rightarrow$ Input Request, $D \rightarrow$ Service Repository

Data: $X \rightarrow$ the sorted list of composition to explore

Output: $C \rightarrow$ the solution composition, or $\emptyset$

1.Begin

2. foreach $outR \in R_{out}$ **do**

3. foreach $s \in D \mid c \in s_o(outR \text{ subsumes } c)$ **do**

4. $X \leftarrow \text{match}(outR, outA);$

5. $C_v \leftarrow \{ s \};$

6. end

7. end

8. while $X \neq \emptyset$ **do**

9. $X \leftarrow \text{sort}(\text{descending}, X, c);$

10. $C \leftarrow \text{popLastElement}(X);$

11. if $\text{isBest}(C)$ **then**

12. return $C;$

13. foreach $c \in \text{unknown}(C)$ **do**

14. foreach $s \in \text{promising}(c)$ **do**

15. $X \leftarrow \text{append}(X, s \oplus C);$

16. end

17. end

18. *end*

19. *return* $\emptyset$

20. *end*

Algorithm Description

- In algorithm, line 1 to line 7 are used to search service by comparing requester output with repository services output and store that service in Cv. That also define function of matchmaking algorithm that call second algorithm of matchmaking conditions.
- Line 8 used to check whether the set X is empty or not. The set X is used to denote set of composition. The set C is used to store final resultant composition.
- Line 9 is used to sort compositions set X according to comparator function C_{cmp} and put them in descending order and store it in X.
- Line 10 is used to pop last composition from composition set X and store it in C.
- in line 11 and line 12 it will check that is C is best composition than it will return that composition.
- from line 13 to line 17 it will take input of composition C and match that input with output of service and services which match are appended to composition set x.
- Line 8 to line 17 are repeated until $x \neq \emptyset$.

Algorithm 2

This algorithm defines the Degree of match for matching the requested services.

Input: outR (required concept), outA (advertised concept)

Output: Exact **or** Plug-in **or** Subsumes **or** Fail

1.	if $outR = outA$ then
2.	return Exact
3.	else if $outA$ subsumes $outR$ then
4.	return Plugin
5.	else if $outR$ subsumes $outA$ then
6.	return Subsumes
7.	Else
8.	return Fail
9	end if Algorithm Description

	Take input as required concept outR and advertised concept outA. Check outR is equal to outA. If yes, then return exact condition.
	If no, then check outA is superclass of outR. If yes, then return plug-in condition.
	If no, then check outR is superclass of outA. If yes, then return subsumes condition.
	If no, then return fail condition.

FLOWCHART

Heuristic composition Algorithm depend on No. of service in repository so as we increases no. of service execution time is also increase. The time complexity in the worst case is $O(b^m)$

Where b is the maximum number successors of any nodes, namely, the number of services that can produce any concept required by the candidate composition, m is the maximum depth of the search space, namely, the maximum number of services in a composition.

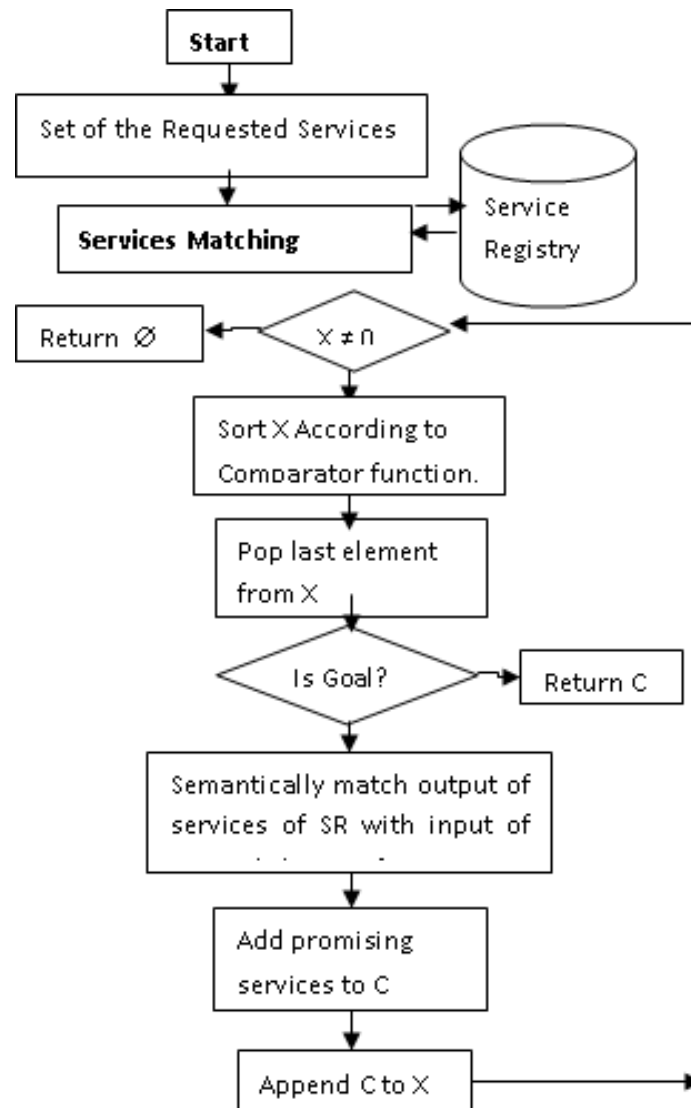


Figure 2: Flowchart of Composition Algorithm

EXPERIMENTAL SETUP

A. Web Service Challenge (WSC) Test Set

- The web service challenge started in 2005 to stimulate research into web service composition, growing and evolving each year.
- During the years 2005 to 2007, the Web Service Challenge focused on optimizing the service discovery and composition process solely, using abstractions from real-world situations. The taxonomies of semantic concepts as well as the involved data formats were purely artificial.
- Starting with the 2008 competition, the data formats and the contest data itself will be based on the OWL, WSDL, and WSPEL schemas for ontologies, services, and service orchestrations to mimic real world scenario.
- In 2009, each service is annotated with nonfunctional properties. The Quality of Service (QoS) of a Web Service is expressed by values expressing its response time and throughput.

B. Hardware Configuration

All experiments were performed on a PC platform with a Intel(R) Core(TM) i3 CPU (2.40 GHz), Windows 7 Home Basic, and 3.00 GB RAM, 64 bit Operating System All algorithms were implemented in Java with the use of the Eclipse tool.

C. Experiment Results with Comparison

This section defines the comparison of different matchmaking algorithms for execution time. This defines the Greedy approach, Bipartiate graph, Brute force method for matching the requested services.

Table 1: Different Matchmaking Methods Results

No of Services	Execution Time (ms)		
	Greedy Approach	Bipartiate Graph Matching	BruteForce Method
10	13	15	15
50	22	49	50
100	29	60	75
150	31	90	120
200	40	118	149
250	49	146	201
300	54	227	302

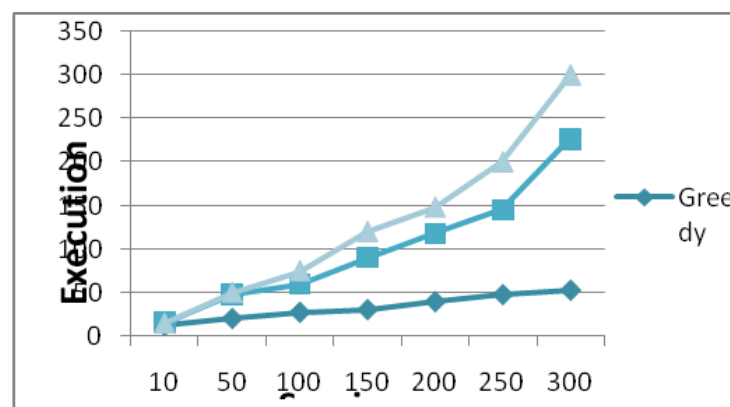


Figure 3: Comparison of No of Services Vs Execution Time

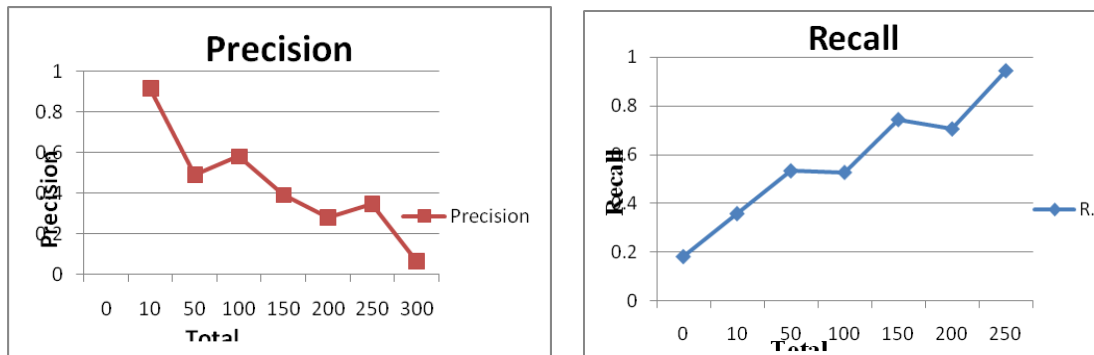


Figure 4: Comparison of No of Services Vs Precision Figure 5: Comparison of No of Services Vs Recall Rate

CONCLUSIONS & FUTURE WORK

We have identified that the heuristic approach also uses the semantic approach and in that we added semantic matchmaking approach for composition of the user requirement services. So semantic matchmaking can be added using the second algorithm that we proposed ahead in proposed work. In this we improved the matchmaking process by the use of the greedy approach. That shows the results that our greedy approach is far better than the other approaches for matching the queries.

To enhance the semantic matchmaking algorithm. To develop a prototype for the proposed work. To study and propose matchmaking algorithm for Cloud Services.

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